

# MARKOV PROCESS

- $(X_n)_{n \geq 0} / (X_t)_{t \geq 0}$  is a MP?  $IP(X_{t_{n+1}} \in E_{n+1} | X_{t_n} \in E_n, \dots, X_{t_0} \in E_0) \stackrel{?}{=} IP(X_{t_{n+1}} \in E_{n+1} | X_{t_n} \in E_n)$

- $(X_n)_{n \geq 0} : X_{n+1} = F(X_n, V_{n+1}) \quad V_{n+1} \text{ def.} \Rightarrow IP(X_{n+1} \in E_{n+1} | X_n = x, X_{n-1} \in E_{n-1}, \dots)$   
if  $= IP(X_{n+1} \in E_{n+1} | X_n = x) \quad \checkmark$

- $(X_n)_{n \geq 0} / (X_t)_{t \geq 0} : X_n = g(B_n) / X_t = g(B_t) \Rightarrow$  use def (\*) and the fact that the Brownian motion is a MP

$$(*) \quad IP(B_{n+1} \in E_{n+1} | B_n = x, B_{n-1} \in E_{n-1}, \dots) = IP(B_{n+1} - B_n + x \in E_{n+1} | B_n - B_{n-1} \in \dots) \\ \stackrel{\text{increments}}{=} IP(B_{n+1} - B_n + x \in E_{n+1}) \\ = IP(B_{n+1} \in E_{n+1} | B_n = x)$$

- Transition Kernel?  $P_t(x, A) = IP(X_t \in A | X_0 = x)$

$$\bullet \quad X_{n+1} = F(X_n, V_{n+1}) \Rightarrow IP(X_{n+1} \in A | X_n = x) = IP(F(x, V_{n+1}) \in A | X_n = x) \\ = IP(F(x, V_{n+1}) \in A)$$

$$Y := F(x, V_{n+1}) ; IP(Y \leq y) = IP(F(x, V_{n+1}) \leq y) = IP(V_{n+1} \leq \dots)$$

$$\text{and we get } F_Y(y) = F_{\dots}(\dots)$$

$$\Rightarrow \frac{\partial F_Y}{\partial y} = f_Y(y) = \frac{\partial F_{\dots}(\dots)}{\partial y} \Rightarrow \text{we obtain } f_Y(y) \text{ using the law of } V_{n+1}$$

$$IP(X_{n+1} \in A | X_n = x) = IP(Y \in A) = \int_A f_Y(y) dy$$

- $X_t = g(B_t) \Rightarrow$  if  $g$  is a linear comb.  $\Rightarrow B_t \sim N(\cdot, \cdot) \Rightarrow X_t \sim N(\cdot, \cdot)$

$$\begin{bmatrix} X_{t+s} \\ X_s \end{bmatrix} = \begin{bmatrix} g(B_{t+s}) \\ g(B_s) \end{bmatrix} \sim N\left(\begin{bmatrix} \mu_Y \\ \mu_X \end{bmatrix}, \begin{bmatrix} \sigma_Y^2 & \text{cov} \\ \text{cov} & \sigma_X^2 \end{bmatrix}\right)$$

$$\boxed{Y | X = x \sim N\left(\mu_Y + \text{Cov}(X, Y) \frac{(x - \mu_X)}{\sigma_X^2}, \sigma_Y^2 - \frac{(\text{Cov}(X, Y))^2}{\sigma_X^2}\right)}$$

$$\Rightarrow X_{t+s} | X_s = x \sim \dots$$

$$\Rightarrow IP(X_{t+s} \in A | X_s = x) = P_t(x, A) = \int_A f_{X_{t+s} | X_s}(y | x) dy$$

$$\boxed{Y = AX, X \sim N(\mu_X, C) \Rightarrow Y \sim (A\mu_X, ACA^T)}$$

- Irreducible?

$$\text{Irreducible} \iff [\forall x \in E, \forall A \in \mathcal{E}, \varphi(A) > 0 \Rightarrow \exists t > 0 : P_t(x, A) > 0]$$

$$\text{In our cases (most generally): } E = \mathbb{R}, \varphi \text{ Leb-meas.} \Rightarrow [P_t(x, A) > 0 \iff \varphi(A) > 0] \\ (\text{once we calculated } P_t(x, A))$$

- Homogeneous?

$$\text{Homogeneous} \iff P_t(x, A) = IP(X_{t+s} \in A | X_s = x) \quad \perp s$$

martingale  $X_n$

$$P_n(X_n, A) = \mathbb{E}[\mathbb{1}_A(X_{n+1}) | \sigma(X_1, \dots, X_n)]$$

$$\stackrel{!}{=} \mathbb{E}[\mathbb{1}_A(h(Z_{n+1}) X_n) | \sigma(\dots)]$$

$$\stackrel{!}{=} \int_{\mathbb{R}} \mathbb{1}_A(h(x) X_n) \cdot f_{Z_{n+1}}(x) dx$$

$$\stackrel{!}{=} \int_{\mathbb{R}} \mathbb{1}_A(y) f_{Z_{n+1}}(\cdot \cdot y) dy \dots$$

$$y = h(x) X_n, \quad dx = \dots dy$$

here  $X_n$  becomes  $x$  and we get  $P_n(x, A)$

# MARTINGALE

• Prove it's a martingale.

1.  $E[|M_t|] < \infty \quad \forall t$

2.  $M_t$  is  $\sigma(N_t)$ -meas.  $(N_t)_{t \geq 0}$  natural filtration

3.  $\forall s \leq t : E[M_t | \sigma(N_s)] = M_s$  (\*\*\*)

3. (desperate trick):  $M_n = f(X_n)$ ,  $\mathcal{O} = \sigma(X_j : j \leq n)$

if  $E[ZW] = E[XW] \quad \forall W \mathcal{O}$ -measurable  $\Rightarrow Z = E[X | \mathcal{O}]$

Here  $X = M_{n+1}$ ,  $Z = M_n$ ,  $W = g(X_0, \dots, X_n)$  generic  $g$ :

$$\begin{aligned} E[M_{n+1} g(X_0, \dots, X_n)] &= \sum_{\delta_0, \dots, \delta_n \in I} f(\delta_{n+1}) g(\delta_0, \dots, \delta_n) P(X_0 = \delta_0, \dots, X_{n+1} = \delta_{n+1}) \\ &= \sum_{\delta_0, \dots, \delta_n \in I} \left( f(\delta_{n+1}) P_{\delta_n} \delta_{n+1} \right) g(\delta_0, \dots, \delta_n) P(X_0 = \delta_0, \dots, X_n = \delta_n) \\ &= [\dots] = E[M_n g(X_0, \dots, X_n)] \end{aligned}$$

• (M/M/k) Starting from  $\mathcal{Q}$ , given  $f(j) = k \quad j \in I$ :

is  $M_t = f(X_t) - \int_0^t \dots \mathbb{1}_{\dots} ds$  a martingale?

\*  $Qf\left(\begin{bmatrix} j \\ \vdots \end{bmatrix}\right) = \begin{bmatrix} \vdots \\ \vdots \end{bmatrix} \stackrel{?}{=} \dots \mathbb{1}_{\dots}$  if so  $\Rightarrow M_t = f(X_t) - \int_0^t (Qf)(X_s) ds$  is a martingale

$\uparrow$   
 $Q \cdot f[\cdot] = [\cdot] \cdot [\cdot]$

In discrete time:  $M_n = f(X_n) - \sum_{k=0}^{n-1} [Pf(X_k) - f(X_k)]$  (\*\*\*)

• (M/M/k) Average service time? Mean time for going back to  $\{0\}$ ?  
(w/ martingale stopping theorem)

$E_1[T_0] = E_1[\inf\{t > 0 : X_t = 0\}]$  = mean time for going  $\{1\} \rightarrow \{0\}$

Stopping theorem:  $E_1[M_0] = E_1[M_{T \wedge t}] = E_1[M_t] \quad t \geq 0$

Since  $T \wedge t \xrightarrow{t \rightarrow \infty} T \Rightarrow$  Monotone conv. theorem:  $E_1[T \wedge t] = E_1[T]$   
 $\Rightarrow$  Lebesgue's theorem:  $E_1[f(X_{T \wedge t})] \rightarrow E_1[f(X_T)]$

$E_1[M_0] = E_1[f(X_0) - \int_0^0 \dots] = E_1[f(X_0)] = f(1)$

$E_1[M_T] = E_1[f(X_T) - \int_0^T \dots] = E_1[f(X_T)] - E[\int_0^T \dots] = f(0) - E[-T]$

$\Rightarrow E_1[M_0] = E_1[M_T] \Rightarrow E[T] = \dots$

(\*\*\*) :  $E[f(X_n)] < \infty$  and  $E[|Pf(X_n)|] < \infty$

\* before: "  $M_t$  is adapted to the natural filtration of the MC "

(\*\*\*) If it's discrete:

"  $E[M_{n+1} | \mathcal{F}_n] = \dots = M_n$ , and so by the projective property of the conditional expectation  $E[M_m | \mathcal{F}_s] = M_s \quad \forall m \geq s$  "

$M_t = f(X_t) - \int_0^t \mathbb{1}_{\{X_s > 0\}} ds \xrightarrow{t \rightarrow \infty} f(X_0) - \int_0^\infty \mathbb{1}_{\{X_s > 0\}} ds$  total time spent in the transient states (since  $\int_0^\infty \mathbb{1}_{\{X_s > 0\}} ds < \infty$  then all the  $M_t$  converges in  $L^1$ )  
if 0 only recurrent

# CONTINUOUS MC (Discrete states space)

- starting from  $Q$  :  $P_t$  ? \*

FKE :  $p_{ij}'(t) = \sum_{k \in I} p_{ik}(t) q_{kj}$

BKE :  $p_{ij}'(t) = \sum_{k \in I} q_{ik} p_{kj}(t)$

and :  $p_{ij}(0) = \delta_{ij} = \mathbb{1}_{\{i=j\}}$  (since  $P_0 = I$  (condition to use to determine  $\alpha$  in  $P_t$ ))

- Starting from  $P_t$  :  $Q$  ?

$$Q = \left. \frac{dP_t}{dt} \right|_{t=0}$$

- Invariant density  $\pi$  ?

$\pi$  invariant  $\iff 0 = \pi Q$   
(Remark :  $I$  finite  $\implies \exists (\pi_j)_j$ )

- Recurent / transient state ?

$i$  is recurrent/transient for the continuous MC  $(X_t)_{t \geq 0}$

$\iff$

$i$  is recurrent/transient for the DISCRETE SKELETON :  $(Y_n)_{n \geq 0}$  :

$$\hat{p}_{ij} = \begin{cases} \frac{q_{ij}}{-q_{ii}} & q_{ii} \neq 0, i \neq j \\ 1 & q_{ii} = 0, i = j \\ 0 & (q_{ii} = 0, i \neq j) \vee (q_{ii} \neq 0, i = j) \end{cases}$$

$q_{ii} = 0 \implies i$  absorbing (recurrent)

- Probability of going to  $j$  after leaving  $i$  ?

If  $q_{ii} \in (-\infty, 0) \implies \forall j \neq i \quad P_i(X_{T_i} = j) = \frac{q_{ij}}{-q_{ii}}$

- Irreducible MC ?

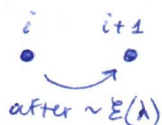
$(X_t)_{t \geq 0}$  irreducible  $\iff \forall i, j \in I : \exists n \geq 1, \exists k_1, \dots, k_n$  s.t.  $i \neq k_1, \dots, k_n \neq j$  and  $q_{ik_1} q_{k_1 k_2} \dots q_{k_n j} > 0$

- Average time spent in  $i$  starting from  $i$  ?

$-\infty < q_{ii} < 0 \implies T_i \sim \mathcal{E}(-q_{ii})$  = exit time from  $i$

$E_i[T_i]$  = average time spent in  $i$  starting from  $i$  =  $\frac{1}{-q_{ii}}$

- POISSON PROCESS



$$Q = \begin{bmatrix} -\lambda & \lambda & 0 & 0 & 0 & \dots \\ 0 & -\lambda & \lambda & 0 & 0 & \dots \\ 0 & 0 & -\lambda & \lambda & 0 & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots \end{bmatrix}, \quad \hat{P} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & \dots \\ 0 & 0 & 1 & 0 & 0 & \dots \\ 0 & 0 & 0 & 1 & 0 & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots \end{bmatrix}$$

$(N_t)_{t \geq 0}$  :  $N_t$  = how many people till time  $t$  :

$N_t \sim P(\lambda t) : P(N_t = k) = \frac{e^{-\lambda t} (\lambda t)^k}{k!}$

$N_{t+s} - N_s \sim P(\lambda t) : P(N(t+s) - N(s) = k) = \frac{e^{-\lambda t} (\lambda t)^k}{k!}$   
 $N_t - N_s \sim P(\lambda(t-s))$

- M/M/K

- H<sub>p</sub> :
- arriving time  $\sim \mathcal{E}(\lambda)$
  - permanence time  $\sim \mathcal{E}(\mu)$
  - $k$  counters (servers)
  - all of them are i.i.d. r.v.
  - $(X_t)_{t \geq 0}$  :  $X_t$  = # customers at the time  $t$

general model

$$Q = \begin{bmatrix} -\lambda & \lambda & 0 & 0 & 0 & \dots \\ \mu & -(\lambda + \mu) & \lambda & 0 & 0 & \dots \\ 0 & \mu & -(\lambda + \mu) & \lambda & 0 & \dots \\ 0 & 0 & \mu & -(\lambda + \mu) & \lambda & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots \end{bmatrix}$$



$T_i = \text{exit time from } i \sim \xi(\cdot) \Rightarrow q_{ik} = \dots(\cdot)$

- $\Rightarrow Q$  generally different

- (Stationarity conditions)

$$= (\text{service time}) + (\text{waiting time if there are already } k \text{ customers})$$

$$\frac{1}{\mu} + \sum_{\substack{j \geq k \\ j \in I}} \boxed{\mathbb{E}_{\pi}[Y_t | X_t = j]} \pi_j$$

if  $X_t = j$ , and we have  $k$  counts:

We have to wait for  $(j-k)+1$  counters to get free:

- Stationary conditions  $\Rightarrow X_t \sim \pi$

Little law:  $E_{\pi}[X_t] = \lambda E_{\pi}[W_t]$

- $n$  counters get free  $\sim \Gamma(n, \underbrace{\frac{1}{\mu} + \dots + \frac{1}{\mu}}_{k \text{ times}})$

- $$\text{arriving} = \text{birth} \sim \mathcal{E}(\lambda) \implies P(B(i) < D(i)) = \frac{\lambda}{\mu + \lambda}$$

$$= P_i(N_{Ti} = j) = \frac{q_{ij}}{-q_{ii}}$$

- probability of  
going to  $j$  after  
leaving  $i$   
(for instance  $j=i+1$ )

- BROWNIAN MOTION

$$B_t \sim N(0, \sigma^2 t)$$

$$B_{t+s} - B_t \sim N(0, s\sigma^2)$$

$$E[(B_{t+s})(B_t)] = \sigma^2 t$$

$$\mathbb{E}[B_t | \mathcal{O}(B_s)] = \mathbb{E}[B_t - B_s] + B_s = B_s$$

$$X_t(B) \perp\!\!\!\perp X_s(B) ? \implies \begin{bmatrix} X_t(B) \\ X_s(B) \end{bmatrix} \sim N(\dots) \text{ and if } \text{Cov}(X_t, X_s) = 0 \implies X_t \perp\!\!\!\perp X_s$$

- due moli, 4 navi
- scarico molo 1  $\sim \mathcal{E}(a)$
- scarico molo 2  $\sim \mathcal{E}(b)$
- tempo tra wi una nave lascia e torna  $\sim \mathcal{E}(\lambda)$

Q:  $X_t = \# \text{navi al molo}$

	0	1	2	3	4	
0	-4λ	4λ				0
1	a	-(a+b)	3λ			1
2		a+b	-(a+b)	2λ		2
3			a+b	-(a+b)	λ	3
4				a+b	-(a+b)	4

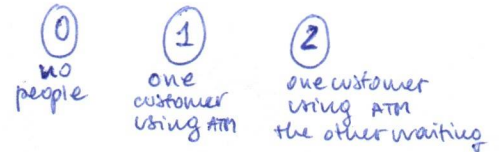
- 2 armchairs: every one must do both of them in order 1+2
- time at 1  $\sim \mathcal{E}(1)$
- time at 2  $\sim \mathcal{E}(1)$
- people go if 1 is empty
- if we're in 1 and done but 2 is occupied then we wait

States = situations:

- 0 = no people
- 1 = one person in the first chair
- 2 = one person in the second
- 3 = both persons (2) served
- 4 = person in 1 waiting, 2 be doing

	0	1	2	3	4	
0	-1/2	1/2				0
1		-1	1			1
2	1		-3/2	1/2		2
3		1		-2	1	3
4			1		-1	4

- one ATM, one place for waiting
- the waiter might get a call ( $\sim \mathcal{E}(\alpha)$ )
- average time service  $\sim \mathcal{E}(\mu)$
- arrivals  $\sim \mathcal{E}(\lambda)$



Fraction of people finishing?

frac. people entering =  $\pi_0 + \pi_1$

Who enters  $\pi_0 \Rightarrow$  finish

Who enters  $\pi_1 \Rightarrow$  finish if the other conclude before one call:

$$P(\text{finish} < \text{call}) = \frac{\mu}{\alpha + \mu}$$

frac. people finishing:

$$\frac{\pi_0}{\pi_1 + \pi_0} + \frac{\pi_1}{\pi_1 + \pi_0} \left( \frac{\mu}{\alpha + \mu} \right)$$

€ c / hour, costs of machines  
€ b for each operation



the bank charges b €/hour whenever the machine is busy:

$$b(\pi_1 + \pi_2) > c \quad \text{for self-maintenance}$$

average profit per hour

- one ATM, max 2 customers
- arrivals  $\sim \mathcal{E}(\lambda)$
- operation  $\sim \mathcal{E}(\mu)$
- second operation  $\sim \mathcal{E}(\alpha)$   
(only p persons do it ( $p = \%$ ))

States = situations:

- 0 = no customers
- 1 = one customer doing his 1st
- 2 = one customer doing his 2nd
- 3 = two customers, one is doing his 1st
- 4 = two customers, one is doing his 2nd

	0	1	2	3	4	
0	-λ	λ				
1	(1-p)μ	-(λ+μ)	pμ	λ		
2	α		-(α+λ)			
3		(1-p)μ		-μ	λ	
4		α			-α	

starting from i the jump to j occurs in an exponential time of param A with prob B  $\Rightarrow q_{ij} = AB$

\* • Frac. of people using ATM?

$$\pi_0 + \pi_1 + \pi_2$$

• Prob. that a customer who ~~need~~ using ATM comes when a customer is doing his 2<sup>nd</sup> operation?

$$\frac{\pi_2}{\pi_0 + \pi_1 + \pi_2}$$

• How much time do we have to wait on average without knowing if the customer is doing 1<sup>st</sup> or 2<sup>nd</sup>?

$$\frac{\pi_1}{\pi_0 + \pi_1 + \pi_2} \cdot T_1 + \frac{\pi_2}{\pi_0 + \pi_1 + \pi_2} T_2$$

probability of using ATM and coming when one is using the ATM for the 1<sup>st</sup> op.

prob. of using the ATM and coming when one is doing the 2<sup>nd</sup>

$T_2$  = mean time for the second op.

$T_1$  = mean time for the first op. and (eventualmente) for the second

$$T_2 = \frac{1}{\lambda} , \quad T_1 = \frac{1}{\mu} + \frac{\rho}{\lambda}$$



# DISCRETE MC

- $(X_n)_{n \geq 0}$  is a MC?

$$IP(X_{n+1}=j | X_n=i, X_{n-1}=i_{n-1}, \dots, X_0=i_0)$$

- If we can, we can calculate

$$\frac{IP(X_{n+1}=j, X_n=i, \dots, X_0=i_0)}{IP(X_n=i, \dots, X_0=i_0)}$$

and see if it depends only on  $i$  and  $j$

- if given  $i$   $X_n$  is defined and  $X_{n+1}$  depends only on  $i$  and  $j$ :  

$$= IP(X_{n+1}=j | X_n=i) \Rightarrow MC$$

$$= p_{ij}$$

- If it's needed go with subtractions:

$$= IP(X_{n+1} - X_n = j - i | X_n = i, \dots)$$

again, if  $X_n$  depends only on  $i \Rightarrow MC$

- if  $z = [-]$   $= IP(Z_{n+1} = (j-i) - (i-i_{n-1}) | S_n = (i-i_{n-1}), \dots)$   
then it depends not only on  $i \Rightarrow$  (probably) not a MC

Counter-example: try something like:  $IP(X_3=3 | X_2=1, X_1=0), IP(X_3=3 | X_2=1, X_1=1)$

I

$$\boxed{\bullet} IP(X_n=k) = \pi_k^{(n)}; \pi^{(0)} = (\dots, IP(X_0=k), \dots), \pi^{(n)} = p^n \pi^{(0)} \quad \begin{matrix} P = QDQ^{-1} \\ p^n = QD^nQ^{-1} \end{matrix}$$

$$\bullet IP(X_n=k, X_m=j) = IP(X_n=k | X_m=j) IP(X_m=j) \stackrel{H}{=} IP(X_{n-m}=k | X_0=j) IP(X_m=j)$$

$$= (p^{n-m})_{jk} \cdot IP(X_m=j)$$

$$\bullet IP(X_n + X_0 = k) = \sum_j IP(X_n + X_0 = k | X_0 = j) IP(X_0 = j) = [-]$$

$$\bullet IP(X_n=i, X_0=j | X_n + X_0 = k) = \frac{IP(X_n + X_0 = k | X_0=j, X_n=i) IP(X_0=j, X_n=i)}{IP(X_n + X_0 = k)}$$

$$\bullet E[X_n X_m] = \sum_{i,j} ij IP(X_n=i, X_m=j)$$

II

$p^n$  not easy

$$\boxed{\bullet} IP(X_n=k) = \sum_j IP(X_n=k | X_0=j) IP(X_0=j)$$

then proceed looking for the possible paths from  $j \rightarrow \dots \rightarrow k$

- Homogeneity? For instance check:  $IP(X_2=j | X_1=i) \stackrel{?}{=} IP(X_1=j | X_0=i)$

- closed class:  $C$  is a closed class if:  $\begin{cases} i \in C \\ j \text{ accessible from } i \end{cases} \Rightarrow j \in C$

$$\bullet \pi? \quad (\pi = \pi P) \wedge (\sum \pi_i = 1)$$

I

- $\boxed{\bullet}$  Hitting prob. of a recurrent state (/class) starting from transient?  
= absorption probability  $(V_i)_{i \in I}$ :

$$V_i = \sum_{j \in C} p_{ij} + \sum_{j \in T} p_{ij} V_j$$

$i$  transient

$C$  = set of all the recurrent states we want to consider

- law of the hitting time of a recurrent state (/class) starting from  $T$ ?

$$T_C = \inf \{n \geq 1 : X_n = k, k \in C\}, \quad X_0 = \bar{i}$$

$$IP(T_C = n) = IP(X_n \in C, X_{n-1} \notin C, \dots, X_1 \notin C | X_0 = \bar{i})$$

$$\downarrow \sum_{j \in C} \sum_{i_{n-1}, \dots, i_1 \notin C} IP(X_n=j | X_{n-1}=i_{n-1}) IP(X_{n-1}=i_{n-1} | X_{n-2}=i_{n-2}) \dots IP(X_1=i_1 | X_0=\bar{i})$$

$$\downarrow \sum_C \sum_{\notin C} p_{i_{n-1}j} p_{i_{n-2}i_{n-1}} \dots p_{\bar{i}i_1}$$

≠ tempo di soggiorno (PDF)

Moreover:  $P(T_c = +\infty) = 1 - P(T_c < +\infty) = 1 - \sum_n P(T_c = n)$

Mean absorption time in  $C$  (recurrent) starting from  $i \in T$ ?

$= W_i \quad i \in T$  :  $W_i = 1 + \sum_{j \in T} p_{ij} W_j \quad i \in T$

stopping time  
 $W_i = E_i[T]$   
(we know that  $E_i[T] = \infty$  if the MC is transient or non recurrent)

I (Irreducible) MC recurrent? Remark: irreducible + finite  $\Rightarrow$  recurrent

$P_0(T_0 < +\infty) = \sum_{n \geq 1} P_0(T_0 = n) = \sum_{n \geq 1} P(X_n = 0, X_{n-1} \neq 0, \dots, X_1 \neq 0 | X_0 = 0)$   
 $= \sum_{n \geq 1} P(X_n = 0 | X_{n-1} \neq 0) P(X_{n-1} \neq 0 | X_{n-2} \neq 0) \dots P(X_1 \neq 0 | X_0 = 0)$

If  $P_0(T_0 < +\infty) = 1 \Rightarrow$  MC recurrent

I<sup>(1)</sup> Existence of invariant distr.  $\pi$ ? (Irreducible MC)

Remark:  
I finite  $\Rightarrow \exists (\pi_j)$

Irreducible MC  $\Rightarrow$  if 0 is recurrent ( $\uparrow$ ) :  $E_0[T_0] = \sum_{n \geq 1} n \cdot P(T_0 = n)$

If  $E_0[T_0] < \infty \Rightarrow$  0 is positive recurrent

$\Rightarrow \exists! \pi$  invariant and  $\pi_0 = \frac{1}{E_0[T_0]}$

I<sup>(2)</sup> Non-existence of inv. distr.  $\pi$ ? (Irreducible MC)

Irreducible MC transient :  $(\pi_j)$  invariant  $\Leftrightarrow \pi P = \pi \Leftrightarrow \pi P^n = \pi$   
 $\Leftrightarrow \begin{cases} \pi = \pi P^n \\ \pi_j = \sum_{i \geq 0} p_{ij}^{(n)} \pi_i \end{cases}$

but:

$j$  transient  $\Rightarrow p_{ij}^{(n)} \xrightarrow{n \rightarrow \infty} 0 \quad \forall i \in I$

$\Rightarrow \pi_j = \sum_{i \geq 0} p_{ij}^{(n)} \pi_i \xrightarrow{n \rightarrow \infty} 0 \quad \forall j$

but  $\sum_j \pi_j = 1$ , contradiction  $\Rightarrow \nexists (\pi_j)$

II if we don't know the law of  $T_0$ :  $P_0(T_0 = n)$  (Irreducible) MC transient? Recurrent? (not transient  $\Rightarrow$  recurrent)

I<sup>\*</sup>: (longest)

Consider the state 0:

$T_0 = \inf \{n \geq 1 : X_n = 0\}$  = hitting time of  $\{0\}$

$V_i = P_i(T_0 < +\infty) \begin{cases} = 1 & i \text{ recurrent} \\ < 1 & i \text{ transient} \end{cases} \Rightarrow \begin{matrix} \text{MC recurrent} \\ \text{MC transient} \end{matrix}$

$V_j = p_{ji} + \sum_{k \in I \setminus \{i\}} p_{jk} V_k$

constraints:

- constraints:
- $0 \leq V_j \leq 1$
  - $V_j$  is minimal
  - boundary cond ( $V_1 = \dots, \dots$ )

Steps:

- $i=0,1: V_0 = \dots, V_1 = \dots$
- $i \geq 1: V_i = \dots$
- $V_i = A + B(\#)^i + C(\#)^i$
- constraints to determine  $A, B, C, \dots$
- $V_0 = ?$

II<sup>\*</sup>: (coolest)

$(X_n)_{n \geq 0}$  transient  $\Leftrightarrow \exists$  bounded, non-const. sol.:  $\sum_{k \in I} p_{jk} y_k = y_j \quad j \in I$   
( $\forall j$ , tranne che al più una)

Steps:

$j \geq 1: y_j = \dots$

$y_j = A + B(\#)^i + C(\#)^j$

general cond. for  $A, B, C, \dots$  to be bounded and non-const

If, for example,  $B$  and  $C$  must be  $= 0$  for boundedness  $\Rightarrow \nexists \text{ sol} \Rightarrow$  recurrent

$\Rightarrow$  we take a random value for  $A, B, C \neq 0$  (usually 1) (A can be 0)



## DISCRETE MC (II)

### II Existence of invariant distr. of $\pi$ ? (Irreducible MC)

Irreducible  $(X_n)_{n \geq 0}$  MC

$\exists (y_j)_j, (x_j)_j$  unbounded:

$$\sum_{k \geq 0} p_{jk} y_k = y_j - x_j \quad \forall j$$

$\Rightarrow \exists! (\pi_j)_j$  invariant

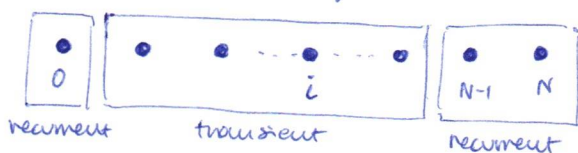
Steps:

- Suppose  $y_j = j \Rightarrow \sum_{k \geq 0} p_{jk} y_k = \sum_{k \geq 0} p_{jk} k = [\cdot] = j - (-)$
- If  $(-)$   $\rightarrow \infty$  we're done
- otherwise suppose  $y_j = j^2 \Rightarrow \dots$

### Absorption probability in a recurrent state (general)? \*

When the equation brings generic  $V_j$  (we can't manually enumerate them)

$$V_i = \sum_{j \in C} p_{ij} + \sum_{j \in T} p_{ij} V_j \quad i \in T$$



Steps:

Suppose we want to be abs. in  $\{0\}$ :

- $V_0 = 1, V_N = V_{N-1} = 0$
- $V_1 = \dots$   
 $V_i = \dots \quad i \geq 2$  ← we need to specify conditions near the abs. class
- $V_i \rightarrow *$
- $V_i = A + B(\#)^i + C(\#)^i \dots$
- Can we write the specials  $(V_0, V_1, V_N, V_{N-1})$  like  $V_i$ ? If so, we impose the conditions to find  $A, B, C, \dots$

if all the MC is transient but one class and the MC is finite  $\Rightarrow$  the prob. of absorption is 1

## \* DIFFERENCE EQUATION

$$X_t = a_1 X_{t-1} + \dots + a_n X_{t-n} + b$$

• **Homogeneous**:  $X_t = a_1 X_{t-1} + \dots + a_n X_{t-n}$

Char. equation:  $\lambda^n = a_1 \lambda^{n-1} + \dots + a_n \Rightarrow \{\lambda_1, \dots, \lambda_n\}$  (\*)

• (all) distinct roots:

$$X_t = c_1 \lambda_1^t + \dots + c_n \lambda_n^t$$

• two equivalent roots:  $\lambda_1 = \lambda_2$ :

$$X_t = c_1 \lambda_1^t + t c_2 \lambda_2^t + \dots$$

• **Non-homogeneous**:  $X_t = a_1 X_{t-1} + \dots + a_n X_{t-n} + b$

Complete solution:  $X_t = (\text{homogeneous solution}) + (\text{particular sol.})$

Particular:  $X_t = D \Rightarrow D = a_1 D + \dots + a_n D + b \Rightarrow D$

$$\Rightarrow X_t = \text{homogeneous} + D$$

(\*) Ex.

$$V_i = \frac{2}{5} V_{i-1} + \frac{2}{5} V_{i+1} \Rightarrow 2V_{i+1} - 5V_i + 2V_{i-1} = 0$$

$$2\lambda^2 - 5\lambda + 2 = 0$$

$$\lambda_{1/2} < \frac{2}{1/2}$$

$$\Rightarrow V_i = A \left(\frac{1}{2}\right)^i + B(2)^i$$

$$W_i = A + B(\#)^i + C(\#)^i \dots \text{ if } \exists \lambda: \lambda = 1$$

# \* DIFFERENTIAL EQUATION

Ex.  $p'_{20}(t) = -\frac{1}{2}p_{20}(t) + \frac{1}{2}e^{-\frac{3}{8}t} - \frac{2}{5}e^{-\frac{5}{12}t} + \frac{2}{5}$  (\*)

Homogeneous:  $p'_{20}(t) = -\frac{1}{2}p_{20}(t) \implies p_{20}(t) = Ke^{-\frac{1}{2}t}$

Complete:  $p_{20}(t) = Ke^{-\frac{1}{2}t} + Ae^{-\frac{3}{8}t} + Be^{-\frac{5}{12}t} + C$

Conditions:  $p'_{20}(t) = \frac{d}{dt}(\uparrow) = (*)$

$p_{20}(0) = \delta_{20} = 0$  (1 if  $\delta_{22}/\delta_{00}/\dots$ )

$$\begin{bmatrix} \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \end{bmatrix}$$

↓  
BKE

Ex.  $p'_{11}(t) = -2p_{11}(t) + p_{12}(t)$  (\*)

Homogeneous:  $p'_{11}(t) = -2p_{11}(t) \implies p_{11}(t) = Ke^{-2t}$

Complete:  $p_{11}(t) = Ke^{-2t} + c(t)e^{-2t}$

Conditions:  $p'_{11}(t) = \frac{d}{dt}(\uparrow) = (*) \implies p_{12}(t) = c'(t)e^{-2t}$   
 $\implies c(t) = \int_0^t p_{12}(s)e^{2s}ds$

$p_{11}(0) = \delta_{11} = 1$

$\implies \underline{p_{11}(t)} = Ke^{-2t} + e^{-2t} \int_0^t p_{12}(s)e^{2s}ds$

$p'_{12}(t) = 1 - 3p_{12}(t)$

↑  
using  $p_{11}(t) + p_{12}(t) + p_{13}(t) = 1$

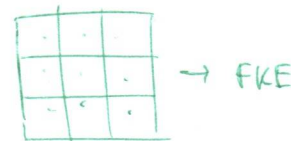
$\implies [\dots \text{Ex. } \uparrow] \implies \underline{p_{12}(t)} = \frac{1}{3}(1 - e^{-3t})$

$\implies \underline{p_{11}(t)} = Ke^{-2t} + e^{-2t} \int_0^t p_{12}(s)e^{2s}ds = [\dots]$

$\implies p_{13}(t) = 1 - p_{11}(t) - p_{12}(t)$

$p'_{00}(t) = -2p_{00}(t) + 2p_{01}(t)$

$p_{00}(t) = e^{-2t} + e^{-2t} \int_0^t e^{2s} p_{01}(s)ds$



$p_{00}(t) = \dots + \dots \int p_{01}(s) \dots ds$

$p'_{01}(t) = p_{00}(t)q_{01} + p_{01}(t)q_{11} + p_{02}(t)q_{21}$

$\oplus p_{00} + p_{01} + p_{02} = 1$

↓  
 $p'_{01}(t) = \dots + p_{01}(t) \dots$

↓  
 $p_{01}(t)$

↓  
 $p_{00}(t)$

↓  
 $p_{02}(t) = 1 - p_{00}(t) - p_{01}(t)$

If  $i, j$  are recurrent in the same class  $C$ :

$\lim_{n \rightarrow \infty} p_{ij}^{(n)} = \pi_j(C)$   
 invariant density restricted on  $C$

moreover for all  $i$  transient and  $j$  recurrent:

$\lim_{n \rightarrow \infty} p_{ij}^{(n)} = u_i(C_j) \pi_j(C_j)$   
 absorption probability in  $C_j$  starting from  $i$

$i$  transient  
 $j$  recurrent

$\lim_{n \rightarrow \infty} p_{ij}^{(n)} = \underbrace{V_i}_{\text{prob. of absorption in the class of } j \text{ starting from } i} \pi_j$

prob. of absorption in the class of  $j$  starting from  $i$